

# CAP 931 – Sales Agent Prototype Documentation

**Student:** Heba Abdelhadi

**Project:** AI Sales Assistant – Account Intelligence

**Duration:** 2 Days

## 1. Technical Setup

This project implements a prototype AI Sales Assistant that generates account intelligence for sales representatives.

### Tech Stack

- Python
- Streamlit (UI)
- FLAN-T5 Large (LLM)
- Requests & BeautifulSoup (web data extraction)
- PyPDF (PDF parsing)

### Files

- app.py: Streamlit UI and input handling
- agents.py: Sales agent logic and prompt orchestration
- utils.py: Web and PDF text extraction
- llm.py: FLAN-T5 summarization logic
- config.py: API keys (excluded from GitHub using .gitignore)

The application collects inputs via Streamlit, extracts public data from URLs and PDFs, summarizes it using an LLM, and outputs a one-page account intelligence report.

## 2. Time Management

Task	Time
<b>Environment &amp; Streamlit setup</b>	<b>2 hours</b>
<b>UI development</b>	<b>3 hours</b>
<b>Web &amp; PDF extraction</b>	<b>4 hours</b>
<b>LLM prompt design</b>	<b>3 hours</b>
<b>Debugging &amp; integration</b>	<b>3 hours</b>
<b>Documentation</b>	<b>1 hour</b>
<b>Total</b>	<b>16 hours</b>

### 3. LLM Model Selection & Use

Model Used: FLAN-T5 Large

Justification:

- Open-source and cost-efficient
- Strong instruction-following and summarization capabilities
- Suitable for structured business insights
- Avoids dependency on paid APIs

Prompt constraints ensure the model only generates sales-related insights.

### 4. Inputs Handling

The Streamlit interface accepts:

- Product Name (required)
- Company URL (required)
- Product Category
- Value Proposition (required)
- Target Customer
- Competitor URLs
- Optional PDF upload

Validation ensures required fields are completed before report generation.

## 5. Data Integration & Output Relevance

- Public web data is extracted from company and competitor URLs
- PDFs are parsed to enrich product context
- Summaries are generated using LLM prompts
- Output is a single-page account intelligence report including:
  - Company Strategy
  - Competitor Mentions
  - Leadership Insights
  - Product Fit Summary
  - Sources

## 6. Challenges and Solutions

**Challenge:** Web scraping restrictions (403 errors)

**Solution:** Used corporate “About” pages and added browser headers.

**Challenge:** Inconsistent LLM responses

**Solution:** Added fallback summaries and minimum-length checks.

**Challenge:** API key security

**Solution:** Stored keys in config.py and added it to .gitignore.

## 7. Experiments

Different prompt formats were tested. Short, role-focused prompts produced the most relevant and concise one-page summaries.

## 8. System Outputs

The generated one-page account intelligence report is included via:

- Screenshots of the Streamlit output

Sources

- <https://corporate.walmart.com/about>

- <https://corporate.target.com/about>
- <https://www.aboutamazon.com/about-us>

## 9. Production Deployment Considerations

For production deployment:

- Host Streamlit on cloud infrastructure
- Use environment variables for secrets
- Add caching, logging, and access controls

## Submission Checklist

Streamlit app runs

One-pager output generated

Screenshots included

Documentation file completed

API keys not committed

GitHub repo clean